

## Chapter 1 : Correlational Designs | educational research techniques

*This lesson explores, with the help of two examples, the basic idea of what a correlation is, the general purpose of using correlational research, and how a researcher might use it in a study.*

In a causal relationship, one variable causes the other. Correlation instead is the association between two variables, which does not necessarily imply causation between them. Experimental designs are aimed at spotting a causal relationship between two variables by isolating them from other control variables factors which may alter the results. Correlation should not be confused with causation: Experimental designs are aimed at spotting a causal relationship between two variables by isolating them from other control variables factors which may alter the results. Correlation designs often are preliminary and less expensive studies which do not yet neutralize control variables, but find that one factor is correlated with another one, meaning that we find these two things together more often than we usually would. We need to find a correlation before we have a causal connection. A statistical generalization is needed in order to establish a correlation, which requires an adequately large sample. Examples can be assessing if there is correlation between smoking and higher rates of heart diseases, or between smoking and gambling. Subsequently, we can ask whether a causal connection is the best explanation for a given correlation. The latter point should not be taken for granted: Let us assume, for instance, that there is an established correlation between smoking X and gambling Y. We have four possible scenarios: X causes Y Something else causes both X and Y family problems, dependent personality traits, other concerns, etc. The correlation is a coincidence A large enough sample is necessary to rule out scenario 4. To assess if there is a causal connection between two variables, it is required that a sample be divided in two groups. The first group, unlike the other, is given the supposed causal factor. For ethical reasons, this form of active test cannot be used if it can potentially cause harm. A passive test can instead be used. This is where the test group consists of cases where we have observed the supposed causal factor to be present, whereas the control group consists of cases where we have observed the causal factor to be absent, all other factors neutralized namely, individuals should match with respect to home background, income level, etc. Social Science Research Design and Statistics:

**Chapter 2 : Quantitative Approaches - Center for Innovation in Research and Teaching**

*A correlation coefficient is usually used during a correlational study. It varies between +1 and -1. A value close to +1 indicates a strong positive correlation while a value close to -1 indicates strong negative correlation.*

Search Introduction to Correlation Research The PowerPoint presentation contains important information for this unit on correlations. Contact the instructor. When are correlation methods used? There is no attempt to manipulate the variables random variables How is correlational research different from experimental research? In correlational research we do not or at least try not to influence any variables but only measure them and look for relations correlations between some set of variables, such as blood pressure and cholesterol level. In experimental research, we manipulate some variables and then measure the effects of this manipulation on other variables; for example, a researcher might artificially increase blood pressure and then record cholesterol level. However, experimental data may potentially provide qualitatively better information: Only experimental data can conclusively demonstrate causal relations between variables. Correlation research asks the question: A correlation has direction and can be either positive or negative note exceptions listed later. With a positive correlation, individuals who score above or below the average mean on one measure tend to score similarly above or below the average on the other measure. The scatterplot of a positive correlation rises from left to right. With negative relationships, an individual who scores above average on one measure tends to score below average on the other or vice versa. The scatterplot of a negative correlation falls from left to right. A correlation can differ in the degree or strength of the relationship with the Pearson product-moment correlation coefficient that relationship is linear. The symbol  $r$  is used to represent the Pearson product-moment correlation coefficient for a sample. The Greek letter rho  $\rho$  is used for a population. When there is no relationship between the measures variables, we say they are unrelated, uncorrelated, orthogonal, or independent. Multiple the  $z$  scores of each pair and add all of those products. Divide that by one less than the number of pairs of scores. Some correlation questions elementary students can investigate are What is the relationship between school attendance and grades in school? Correlations only describe the relationship, they do not prove cause and effect. Correlation is a necessary, but not a sufficient condition for determining causality. There are Three Requirements to Infer a Causal Relationship A statistically significant relationship between the variables The causal variable occurred prior to the other variable There are no other factors that could account for the cause Correlation studies do not meet the last requirement and may not meet the second requirement. However, not having a relationship does mean that one variable did not cause the other. There is a strong relationship between the number of ice cream cones sold and the number of people who drown each month. Just because there is a relationship strong correlation does not mean that one caused the other. Format for correlations research questions and hypotheses: Is there a statistically significant relationship between height and arm span? There is no statistically significant relationship between height and arm span  $H_0$ : There is a statistically significant relationship between height and arm span  $H_A$ : Coefficient of Determination Shared Variation One way researchers often express the strength of the relationship between two variables is by squaring their correlation coefficient. The coefficient of determination is useful because it gives the proportion of the variance of one variable that is predictable from the other variable. Linear Nonlinear or Curvilinear Non-monotonic concave or cyclical. Different procedures are used to measure different types of relationships using different types of scales. Be sure that you understand them. Reading a Correlations Table in a Journal Article Most research studies report the correlations among a set of variables. The results are presented in a table such as the one shown below. The intersection of a row and column shows the correlation between the variable listed for the row and the variable listed for the column. For example, the intersection of the row mathematics and the column science shows that the correlation between mathematics and science was. Most tables do not report the perfect correlation along the diagonal that occurs when a variable is correlated with itself. In the example above, the diagonal was used to report the correlation of the four factors with a different variable. Because the correlation between reading and mathematics can be determined in the top section of the table, the correlations between those two variables is not repeated in the bottom half of the table.

This is true for all of the relationships reported in the table.

**Chapter 3 : Correlation and dependence - Wikipedia**

*1 Robert S Michael Correlation & Ex Post Facto designs-1 Overview: Correlation and Correlational Designs Y Strategies for Educational Inquiry Robert S Michael Robert S Michael Correlation & Ex Post Facto designs*

Each cluster is interpreted as representing a different underlying variable or factor.. In essence, factor analysis organizes the variables into a smaller number of clusters, such that they are strongly correlated within each cluster but weakly correlated between clusters. Each cluster is then interpreted as multiple measures of the same underlying construct. For example, measures of warmth, gregariousness, activity level, and positive emotions tend to be highly correlated with each other and are interpreted as representing the construct of extroversion. The structure and personality correlates of music preferences. *Journal of Personality and Social Psychology*, 84, " They then submitted these 14 variables to a factor analysis, which identified four distinct factors. Two additional points about factor analysis are worth making here. One is that factors are not categories. Instead, factors are constructs that operate independently of each other. So people who are high in extroversion might be high or low in conscientiousness, and people who like reflective and complex music might or might not also like intense and rebellious music. The second point is that factor analysis reveals only the underlying structure of the variables. It is up to researchers to interpret and label the factors and to explain the origin of that particular factor structure. *Behavioral genetics 5th ed. Exploring Causal Relationships*

Another important use of complex correlational research is to explore possible causal relationships among variables. Complex correlational research, however, can often be used to rule out other plausible interpretations. The primary way of doing this is through the statistical control In complex correlational research, accounting for third variables by measuring them and including them in the analysis. Instead of controlling these variables by random assignment or by holding them constant as in an experiment, the researcher measures them and includes them in the statistical analysis. *Having less, giving more: The influence of social class on prosocial behavior. Journal of Personality and Social Psychology*, 99, " In reality, there was no other participant. As these researchers expected, participants who were lower in SES tended to give away more of their points than participants who were higher in SES. This is consistent with the idea that being lower in SES causes people to be more generous. But there are also plausible third variables that could explain this relationship. It could be, for example, that people who are lower in SES tend to be more religious and that it is their greater religiosity that causes them to be more generous. Or it could be that people who are lower in SES tend to come from ethnic groups that emphasize generosity more than other ethnic groups. The researchers dealt with these potential third variables, however, by measuring them and including them in their statistical analyses. They found that neither religiosity nor ethnicity was correlated with generosity and were therefore able to rule them out as third variables. This does not prove that SES causes greater generosity because there could still be other third variables that the researchers did not measure. But by ruling out some of the most plausible third variables, the researchers made a stronger case for SES as the cause of the greater generosity. Many studies of this type use a statistical technique called multiple regression A statistical technique that describes the relationship between multiple independent variables and a single dependent variable in terms of an equation that shows the separate contribution of each independent variable to the dependent variable.. This involves measuring several independent variables  $X_1, X_2, X_3, \dots, X_i$ , all of which are possible causes of a single dependent variable  $Y$ . The result of a multiple regression analysis is an equation that expresses the dependent variable as an additive combination of the independent variables. This regression equation has the following general form: The quantities  $b_1, b_2$ , and so on are regression weights that indicate how large a contribution an independent variable makes, on average, to the dependent variable. Specifically, they indicate how much the dependent variable changes for each one-unit change in the independent variable. The advantage of multiple regression is that it can show whether an independent variable makes a contribution to a dependent variable over and above the contributions made by other independent variables. As a hypothetical example, imagine that a researcher wants to know how the independent variables of income and health relate to the dependent variable of happiness. This is tricky

because income and health are themselves related to each other. Thus if people with greater incomes tend to be happier, then perhaps this is only because they tend to be healthier. Likewise, if people who are healthier tend to be happier, perhaps this is only because they tend to make more money. But a multiple regression analysis including both income and happiness as independent variables would show whether each one makes a contribution to happiness when the other is taken into account. Research like this, by the way, has shown both income and health make extremely small contributions to happiness except in the case of severe poverty or illness; Diener, The science of happiness, and a proposal for a national index. *American Psychologist*, 55, 34â€”

The examples discussed in this section only scratch the surface of how researchers use complex correlational research to explore possible causal relationships among variables. It is important to keep in mind, however, that purely correlational approaches cannot unambiguously establish that one variable causes another. The best they can do is show patterns of relationships that are consistent with some causal interpretations and inconsistent with others.

**Key Takeaways** Researchers often use complex correlational research to explore relationships among several variables in the same study. Complex correlational research can be used to explore possible causal relationships among variables using techniques such as multiple regression. Such designs can show patterns of relationships that are consistent with some causal interpretations and inconsistent with others, but they cannot unambiguously establish that one variable causes another. Make a correlation matrix for a hypothetical study including the variables of depression, anxiety, self-esteem, and happiness. A multiple regression analysis shows that intelligence is not related to performance in the class but that the need for cognition is. Explain what this study has shown in terms of what causes good performance in the critical-thinking course.

## Chapter 4 : Research Designs | Noba

*The strength of the correlation- the stronger the correlation, the closer it is to or How does regression analysis help us to predict? By using a regression line, which summarizes the points on the scatterplot in the best way.*

Rank correlation coefficients[ edit ] Main articles: If, as the one variable increases, the other decreases, the rank correlation coefficients will be negative. However, this view has little mathematical basis, as rank correlation coefficients measure a different type of relationship than the Pearson product-moment correlation coefficient , and are best seen as measures of a different type of association, rather than as alternative measure of the population correlation coefficient. As we go from each pair to the next pair  $x$  increases, and so does  $y$ . Other measures of dependence among random variables[ edit ] See also: The Randomized Dependence Coefficient [12] is a computationally efficient, copula -based measure of dependence between multivariate random variables. RDC is invariant with respect to non-linear scalings of random variables, is capable of discovering a wide range of functional association patterns and takes value zero at independence. The correlation ratio is able to detect almost any functional dependency,[ citation needed ][ clarification needed ] and the entropy -based mutual information , total correlation and dual total correlation are capable of detecting even more general dependencies. These are sometimes referred to as multi-moment correlation measures,[ citation needed ] in comparison to those that consider only second moment pairwise or quadratic dependence. The polychoric correlation is another correlation applied to ordinal data that aims to estimate the correlation between theorised latent variables. One way to capture a more complete view of dependence structure is to consider a copula between them. The coefficient of determination generalizes the correlation coefficient for relationships beyond simple linear regression to multiple regression. Sensitivity to the data distribution[ edit ] Further information: This is true of some correlation statistics as well as their population analogues. Most correlation measures are sensitive to the manner in which  $X$  and  $Y$  are sampled. Dependencies tend to be stronger if viewed over a wider range of values. For example, the Pearson correlation coefficient is defined in terms of moments , and hence will be undefined if the moments are undefined. Measures of dependence based on quantiles are always defined. Sample-based statistics intended to estimate population measures of dependence may or may not have desirable statistical properties such as being unbiased , or asymptotically consistent , based on the spatial structure of the population from which the data were sampled. Sensitivity to the data distribution can be used to an advantage. For example, scaled correlation is designed to use the sensitivity to the range in order to pick out correlations between fast components of time series.

## Chapter 5 : A Correlational Study Tries to Find a Relationship Between Variables

*Learning Objectives. Explain some reasons that researchers use complex correlational designs. Create and interpret a correlation matrix. Describe how researchers can use correlational research to explore causal relationships among variables including the limits of this approach.*

In this module, the four approaches to quantitative research are described and examples are provided. List and explain the four approaches to quantitative research. Provide an example of each method. Describe how to identify the appropriate approach for a particular research problem. There are four main types of quantitative research designs: The differences between the four types primarily relates to the degree the researcher designs for control of the variables in the experiment. Following is a brief description of each type of quantitative research design, as well as chart comparing and contrasting the approaches. A Descriptive Design seeks to describe the current status of a variable or phenomenon. The researcher does not begin with a hypothesis, but typically develops one after the data is collected. Data collection is mostly observational in nature. A Correlational Design explores the relationship between variables using statistical analyses. However, it does not look for cause and effect and therefore, is also mostly observational in terms of data collection. A Quasi-Experimental Design often referred to as Causal-Comparative seeks to establish a cause-effect relationship between two or more variables. The researcher does not assign groups and does not manipulate the independent variable. Control groups are identified and exposed to the variable. Results are compared with results from groups not exposed to the variable. Experimental Designs, often called true experimentation, use the scientific method to establish cause-effect relationship among a group of variables in a research study. Researchers make an effort to control for all variables except the one being manipulated the independent variable. The effects of the independent variable on the dependent variable are collected and analyzed for a relationship. When deciding on the appropriate approach, the Decision Tree from Ebling Library may be helpful. The following video, Quantitative Research Designs, further describes the differences between quantitative research approaches and offers tips on how to decide on methodology. Qualitative and quantitative approaches. Qualitative, quantitative, and mixed methods approaches. Research methods in education and psychology: Introduction to social research: Quantitative and qualitative approaches.

**Chapter 6 : Types of Correlational Studies & Data Analysis - Center for Innovation in Research and Teaching**

*Design Introduction and Focus - Correlational research design can be relational (leading to correlation analysis) and predictive (leading to regression analysis). Correlational (relational) research design is used in those cases when there is an interest to identify the existence, strength and direction of relationships between two variables.*

Articulate the difference between correlational and experimental designs. Understand how to interpret correlations. Understand how experiments help us to infer causality. Understand how surveys relate to correlational and experimental research. Explain what a longitudinal study is. List a strength and weakness of different research designs. In fact, if you were in the audience, you would have likely believed he had psychic powers. Everything looked authenticâ€”this man had to have paranormal abilities! So, why have you probably never heard of him before? Because when Uri was asked to perform his miracles in line with scientific experimentation, he was no longer able to do them. That is, even though it seemed like he was doing the impossible, when he was tested by science, he proved to be nothing more than a clever magician. When we look at dinosaur bones to make educated guesses about extinct life, or systematically chart the heavens to learn about the relationships between stars and planets, or study magicians to figure out how they perform their tricks, we are forming observationsâ€”the foundation of science. Science is the result of systematic and intentional study of the natural world. And psychology is no different. However, there are many ways to test hypotheses in psychological research. Which method you choose will depend on the type of questions you are asking, as well as what resources are available to you. All methods have limitations, which is why the best research uses a variety of methods. Most psychological research can be divided into two types: Which option do you think would bring you the most happiness? At the Corner Perk Cafe customers routinely pay for the drinks of strangers. Is this the way to get the most happiness out of a cup of coffee? The Island Packet, <https://www.islandpacket.com>: Some of the participants were told they must spend the money on themselves, and some were told they must spend the money on others either charity or a gift for someone. Psychologists measure many abstract concepts, such as happiness and intelligence, by beginning with operational definitions of the concepts. See the Noba modules on Intelligence [ <http://www.noba.edu>]. The dependent variable is the variable that is not manipulated at all, or the one where the effect happens. Thus, any observed changes or group differences in happiness can be attributed to whom the money was spent on. What Dunn and her colleagues found was that, after all the spending had been done, the people who had spent the money on others were happier than those who had spent the money on themselves. In other words, spending on others causes us to be happier than spending on ourselves. Do you find this surprising? Or what if some people dropped their toast that morning and it fell jam-side down and ruined their whole day? The most important thing about experiments is random assignment. The experimenter assigns them to a particular condition based on the flip of a coin or the roll of a die or any other random method. Why do researchers do this? But another equally important reason is that random assignment makes it so the groups, on average, are similar on all characteristics except what the experimenter manipulates. By randomly assigning people to conditions self-spending versus other-spending, some people with happy childhoods should end up in each condition. Likewise, some people who had dropped their toast that morning or experienced some other disappointment should end up in each condition. As a result, the distribution of all these factors will generally be consistent across the two groups, and this means that on average the two groups will be relatively equivalent on all these factors. Random assignment is critical to experimentation because if the only difference between the two groups is the independent variable, we can infer that the independent variable is the cause of any observable difference. The class is to be divided evenly between the two teams. If you get to pick the players for your team first, whom will you pick? As a result, your team will be taller and more athletic than the other team. But what if we want the teams to be fair? How can we do this when we have people of varying height and ability? All we have to do is randomly assign players to the two teams. Most likely, some tall and some short people will end up on your team, and some tall and some short people will end up on the other team. The average height of the teams will be approximately the same. That is the power of random assignment! Other considerations In addition to using random assignment, you should avoid

introducing confounds into your experiments. Confounds are things that could undermine your ability to draw causal inferences. For example, if you wanted to test if a new happy pill will make people happier, you could randomly assign participants to take the happy pill or not the independent variable and compare these two groups on their self-reported happiness the dependent variable. However, if some participants know they are getting the happy pill, they might develop expectations that influence their self-reported happiness. This is sometimes known as a placebo effect. Sometimes a person just knowing that he or she is receiving special treatment or something new is enough to actually cause changes in behavior or perception: A related idea is participant demand. This occurs when participants try to behave in a way they think the experimenter wants them to behave. Placebo effects and participant demand often occur unintentionally. Even experimenter expectations can influence the outcome of a study. One way to prevent these confounds from affecting the results of a study is to use a double-blind procedure. In a double-blind procedure, neither the participant nor the experimenter knows which condition the participant is in. At the end of the day, the only difference between groups will be which pills the participants received, allowing the researcher to determine if the happy pill actually caused people to be happier.

**Correlational Designs** When scientists passively observe and measure phenomena it is called correlational research. Here, we do not intervene and change behavior, as we do in experiments. In correlational research, we identify patterns of relationships, but we usually cannot infer what causes what. Importantly, with correlational research, you can examine only two variables at a time, no more and no less. You could use a correlational design which is exactly what Professor Dunn did, too. She asked people how much of their income they spent on others or donated to charity, and later she asked them how happy they were. Do you think these two variables were related? The more money people reported spending on others, the happier they were. More details about the correlation To find out how well two variables correspond, we can plot the relation between the two scores on what is known as a scatterplot Figure 1. In the scatterplot, each dot represents a data point. Importantly, each dot provides us with two pieces of information in this case, information about how good the person rated the past month x-axis and how happy the person felt in the past month y-axis. Which variable is plotted on which axis does not matter. Each dot represents an individual. The association between two variables can be summarized statistically using the correlation coefficient abbreviated as  $r$ . A correlation coefficient provides information about the direction and strength of the association between two variables. For the example above, the direction of the association is positive. This means that people who perceived the past month as being good reported feeling more happy, whereas people who perceived the month as being bad reported feeling less happy. With a positive correlation, the two variables go up or down together. In a scatterplot, the dots form a pattern that extends from the bottom left to the upper right just as they do in Figure 1. The  $r$  value for a positive correlation is indicated by a positive number although, the positive sign is usually omitted. Here, the  $r$  value is .7. A negative correlation is one in which the two variables move in opposite directions. That is, as one variable goes up, the other goes down. Figure 2 shows the association between the average height of males in a country y-axis and the pathogen prevalence or commonness of disease; x-axis of that country. In this scatterplot, each dot represents a country. Notice how the dots extend from the top left to the bottom right. What does this mean in real-world terms? It means that people are shorter in parts of the world where there is more disease. The  $r$  value for a negative correlation is indicated by a negative number that is, it has a minus sign in front of it. Here, it is  $-0.6$ . Each dot represents a country. Chiao, The strength of a correlation has to do with how well the two variables align. The more money people reported spending on others, the happier they reported to be. At this point you may be thinking to yourself, I know a very generous person who gave away lots of money to other people but is miserable! Or maybe you know of a very stingy person who is happy as can be. Yes, there might be exceptions. If an association has many exceptions, it is considered a weak correlation. If an association has few or no exceptions, it is considered a strong correlation.

*Correlation means association - more precisely it is a measure of the extent to which two variables are related. If an increase in one variable tends to be associated with an increase in the other then this is known as a positive correlation.*

Correlational Design Correlational Design Design Introduction and Focus “ Correlational research design can be relational leading to correlation analysis and predictive leading to regression analysis. Correlational relational research design is used in those cases when there is an interest to identify the existence, strength and direction of relationships between two variables. This association cannot be used to draw conclusions with regard to cause-effect relationship between the variables. Does this mean that making more money makes you less smart or that if you do well on tests you will make less money? The answer is no. And in order to interpret the results of the analysis we need to know the context. When do we use the design? Correlational research is supported by relational theories that attempt to test relationships between dimensions or characteristics of individuals, groups or situations or events. These theories explain how phenomena, or their parts are related to one another. The theory about the relationship between the constructs was first introduced by Karl Pearson an English statistician and then expanded by Charles Spearman that developed a method to compute correlation for ranked data Salkind In general terms this type of research seems an answer to a question such as: To what extent do two or more characteristics tend to occur together? Specific Characteristics “ Correlational design is the umbrella terms and there are multiple correlational designs under it. For this reason, it is impossible to make one statement about their specific characteristics. Hence, the most general statement that can be made about the specific characteristics of this design is that this design allows to identify relationships or predictive relationships between variables. The characteristics of different correlational designs need to be carefully studied and discussed in relation to each of them individually. A number of suggestions have been made for sample size. The results of the regression analysis are impacted by sample size for instance. Some of those suggestions did not take into consideration the effect size medium effect size in social science and statistical power. Sampling Method “ Both random and non-random sampling procedures can be used with correlational designs. Data Collection “ Data are usually collected by self-reported measures. The data collection instruments are usually multiple-choice Likert-scale questionnaires that allow to collect interval data. However, secondary data such as student exam scores or some other data collected in the past can also be used in correlational design. Data Analysis “ In relational correlational design, although several variables can be entered into the analysis, a bivariate analysis between two variables is performed, and in the case of multiple variables the output will present bivariate relationships between any two variables entered into the analysis. In predictive correlational design which sometimes also is called regression design, and if more than one independent variable “ multiple regression design , the possible predictive relationship between the outcome and the predictor s is identified. The Pearson product-moment is used to determine the direction and strength of the correlation.

**Chapter 8 : Complex Correlational Designs**

*A correlation coefficient of 0 indicates no correlation. Limitations of Correlational Studies While correlational research can suggest that there is a relationship between two variables, it cannot prove that one variable causes a change in another variable.*

Saul McLeod, updated Correlation means association - more precisely it is a measure of the extent to which two variables are related. There are three possible results of a correlational study: A positive correlation is a relationship between two variables in which both variables either increase or decrease at the same time. An example would be height and weight. Taller people tend to be heavier. A negative correlation is a relationship between two variables in which an increase in one variable is associated with a decrease in the other. An example would be height above sea level and temperature. As you climb the mountain increase in height it gets colder decrease in temperature. A zero correlation exists when there is no relationship between two variables. For example there is no relationship between the amount of tea drunk and level of intelligence. A correlation can be expressed visually. This is done by drawing a scattergram - that is one can plot the figures for one variable against the figures for the other on a graph. Remember, in correlations we are always dealing with paired scores, so the values of the 2 variables taken together will be used to make the diagram. Decide which variable goes on each axis and then simply put a cross at the point where the 2 values coincide. Some uses of Correlations Prediction If there is a relationship between two variables, we can make predictions about one from another. Validity Concurrent validity correlation between a new measure and an established measure. Reliability Test-retest reliability are measures consistent. Inter-rater reliability are observers consistent. Theory verification Predictive validity. The correlation coefficient  $r$  indicates the extent to which the pairs of numbers for these two variables lie on a straight line. Values over zero indicate a positive correlation, while values under zero indicate a negative correlation. Differences between Experiments and Correlations An experiment isolates and manipulates the independent variable to observe its effect on the dependent variable, and controls the environment in order that extraneous variables may be eliminated. Experiments establish cause and effect. A correlation identifies variables and looks for a relationship between them. An experiment tests the effect that an independent variable has upon a dependent variable but a correlation looks for a relationship between two variables. This means that the experiment can predict cause and effect causation but a correlation can only predict a relationship, as another extraneous variable may be involved that it not known about. Strengths of Correlations 1. Correlation allows the researcher to investigate naturally occurring variables that maybe unethical or impractical to test experimentally. For example, it would be unethical to conduct an experiment on whether smoking causes lung cancer. Correlation allows the researcher to clearly and easily see if there is a relationship between variables. This can then be displayed in a graphical form. Limitations of Correlations 1. Correlation is not and cannot be taken to imply causation. Even if there is a very strong association between two variables we cannot assume that one causes the other. For example suppose we found a positive correlation between watching violence on T. It could be that the cause of both these is a third extraneous variable - say for example, growing up in a violent home - and that both the watching of T. Correlation does not allow us to go beyond the data that is given. It would not be legitimate to infer from this that spending 6 hours on homework would be likely to generate 12 G. How to reference this article:

## Chapter 9 : Correlational Designs

*A nonexperimental research design used to describe an individual or a group by having participants complete a survey or questionnaire is called the survey research design. A survey, which is a common measurement tool in the behavioral sciences, is a series of.*

This information can be used either to explain a phenomenon or to make predictions. This post will explain the two forms of correlational design as well as the characteristics of correlational design in general.

**Explanatory Design** An explanatory design seeks to determine to what extent two or more variables co-vary. Co-vary simply means the strength of the relationship of one variable to another. In general, two or more variables can have a strong, weak, or no relationship. This is determined by the product moment correlation coefficient, which is usually referred to as  $r$ . The  $r$  is measured on a scale of -1 to 1. The higher the absolute value the stronger the relationship. Exercise is the explanatory variable and health is the response variable. This means that we are hoping the exercise will explain health or you can say we are hoping that health responds to exercise. This literally means that when exercise goes up one unit, that health improves by 0. In other words, when one increases the other increase as well. Exercise is able to explain a certain amount of the variance amount of change in health. This is done by squaring the  $r$  to get the  $r$ -squared. The higher the  $r$ -squared to more appropriate the model is in explaining the relationship between the explanatory and response variable. This is where regression comes from. This also holds true for a negative relationship but in negative relationships when the explanatory variables increase the response variable decreases. This means that when exercise increases one unit age decreases 0. In other words, more exercises mean that the person is probably younger. In this example, the relationship is strong but indicates that the variables move in opposite directions.

**Prediction Design** Prediction design has most of the same functions as explanatory design with a few minor changes. In prediction design, we normally do not use the term explanatory and response variable. Rather we have predictor and outcome variable as terms. This is because we are trying to predict and not explain. In research, there are many terms for independent and dependent variable and this is because different designs often use different terms. Another difference is the prediction designs are focused on determining future results or forecasting. What both designs have in common is the use of  $r$  and  $r$  square and the analysis of the strength of the relationship among the variables.

**Conclusion** In research, explanatory and prediction correlational designs have a place in understanding data. Which to use depends on the goals of the research and or the research questions. Both designs have Advertisements.